

Appendix - YouGov Survey

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September 16, 2024

```
## Clean the working environment and set up the working directory
```

```
rm(list = ls())
```

```
setwd("/Users/qingwang/Downloads/Data Replication")
```

```
# load the libraries
```

```
library(tidyverse)
```

```
## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
```

```
## v dplyr      1.1.4      v readr      2.1.5
```

```
## v forcats   1.0.0      v stringr    1.5.1
```

```
## v ggplot2   3.5.1      v tibble     3.2.1
```

```
## v lubridate 1.9.3      v tidyr      1.3.1
```

```
## v purrr     1.0.2
```

```
## -- Conflicts ----- tidyverse_conflicts() --
```

```
## x dplyr::filter() masks stats::filter()
```

```
## x dplyr::lag()     masks stats::lag()
```

```
## i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become warnings
```

```
library(kableExtra)
```

```
##
```

```
## Attaching package: 'kableExtra'
```

```
##
```

```
## The following object is masked from 'package:dplyr':
```

```
##
```

```
##      group_rows
```

```
library(haven)
```

```
library(ggthemes)
```

```
#install the flexpath package
```

```
# devtools::install_github("dustinfife/flexplot")
```

```
library(flexplot)
```

```
##
```

```
## Attaching package: 'flexplot'
```

```
##
```

```
## The following object is masked from 'package:ggplot2':
```

```
##
```

```

## flip_data
library(sandwich)
library(lmtest)

## Loading required package: zoo
##
## Attaching package: 'zoo'
##
## The following objects are masked from 'package:base':
##
## as.Date, as.Date.numeric
library(texreg)

## Version: 1.39.4
## Date: 2024-07-23
## Author: Philip Leifeld (University of Manchester)
##
## Consider submitting praise using the praise or praise_interactive functions.
## Please cite the JSS article in your publications -- see citation("texreg").
##
## Attaching package: 'texreg'
##
## The following object is masked from 'package:tidyr':
##
## extract
library(boot)
library(xtable)
library(modelsummary)

## `modelsummary` 2.0.0 now uses `tinytable` as its default table-drawing
## backend. Learn more at: https://vincentarelbundock.github.io/tinytable/
##
## Revert to `kableExtra` for one session:
##
## options(modelsummary_factory_default = 'kableExtra')
## options(modelsummary_factory_latex = 'kableExtra')
## options(modelsummary_factory_html = 'kableExtra')
##
## Silence this message forever:
##
## config_modelsummary(startup_message = FALSE)
library(marginaleffects)
library(ggpubr)
library(pBrackets)
library(lemon)
library(arm)

```

```
## Loading required package: MASS
##
## Attaching package: 'MASS'
##
## The following object is masked from 'package:dplyr':
##
##   select
##
## Loading required package: Matrix
##
## Attaching package: 'Matrix'
##
## The following objects are masked from 'package:tidyr':
##
##   expand, pack, unpack
##
## Loading required package: lme4
##
## arm (Version 1.14-4, built: 2024-4-1)
##
## Working directory is /Users/qingwang/Downloads/Data Replication
##
##
## Attaching package: 'arm'
##
## The following object is masked from 'package:xtable':
##
##   display
##
## The following object is masked from 'package:boot':
##
##   logit
##
## The following object is masked from 'package:flexplot':
##
##   rescale
```

```
library(gplots)
```

```
##
## Attaching package: 'gplots'
##
## The following object is masked from 'package:stats':
##
##   lowess
```

```
library(extrafont)
```

```
## Registering fonts with R
```

```

# install the compactr package
# url <- "https://cran.r-project.org/src/contrib/Archive/compactr/compactr_0.1.tar.gz"
# install.packages(url, repos = NULL, type = "source")
library(compactr)

# load the dataset
df_ces <- read_dta("YouGov/YouGov_clean.dta")
df_ces_post <- read_dta("YouGov/YouGov_post_clean.dta")

```

Table S8: Sample Demographics in Comparison with Census Benchmarks (YouGov Survey)

```

# benchmark demographic data is obtained from Table S0101 of the 2021 American Community Survey
# link to the survey: https://data.census.gov/table/ACSST1Y2021.S0101?q=S0101

```

```

# calculate the demographic of the PureSpectrum sample

```

```

# sex
male_percent <- df_ces %>%
  group_by(male) %>%
  summarise(percentage = round(n() / nrow(df_ces) * 100, 1))
male_percent

```

```

## # A tibble: 2 x 2
##   male percentage
##   <dbl>       <dbl>
## 1     0         52.9
## 2     1         47.1

```

```

# age
# age ==1, 18-29
# age ==2, 30-39
# age ==3, 40-49
# age ==4, 50-59
# age ==5, 60-69
# age ==6, 70 or above

```

```

age_percent <- df_ces %>%
  mutate(age_demo = case_when(age <= 29 ~ '1',
                              age >= 30 & age <= 39 ~ '2',
                              age >= 40 & age <= 49 ~ '3',
                              age >= 50 & age <= 59 ~ '4',
                              age >= 60 & age <= 69 ~ '5',
                              age >= 70 ~ '6')) %>%
  group_by(age_demo) %>%
  summarise(percentage = round(n() / nrow(df_ces) * 100, 1))
age_percent

```

```

## # A tibble: 6 x 2
##   age_demo percentage

```

```
## <chr> <dbl>
## 1 1 15.6
## 2 2 15.2
## 3 3 13.8
## 4 4 18.7
## 5 5 21.4
## 6 6 15.3
```

```
# race
# race ==1, White
# race ==2, Black
# race ==3, Hispanic
# race ==4, Asian
# race ==5, Native American
# race ==6, Two or more races
# race ==7, Other
# race ==8, Middle Eastern

race_percent <- df_ces %>%
  group_by(race) %>%
  summarise(percentage = round(n() / nrow(df_ces) * 100, 1))
race_percent
```

```
## # A tibble: 8 x 2
##   race percentage
##   <dbl> <dbl>
## 1     1    70.7
## 2     2    13.1
## 3     3     8.6
## 4     4     2.7
## 5     5     0.8
## 6     6     2
## 7     7     1.9
## 8     8     0.2
```

```
# calculate the "Other" race category
race_other = race_percent[4,2] + race_percent[5,2] + race_percent[6,2] + race_percent[7,2] + r
race_other
```

```
## percentage
## 1 7.6
```

Table S9: Regression Estimates of Support for War (Binary Dependent Variable, YouGov Survey)

```
# convert the inc variable unit ($ to 10k$)
df_ces <- df_ces %>%
  mutate(inc_10k = inc/10000)

# run the regression models
m1_ces <- lm(attack ~ hmrts + alliance, data = df_ces)
```

Table 1: Linear Estimate of Public Approval for Attacking the Third Country (CES Survey)

	Model 1	Model 2	Model 3
Violating Human Rights	15.71*** (3.03)	20.58*** (4.27)	19.83*** (4.46)
U.S. Military Alliance	-2.78 (3.03)	2.06 (4.26)	2.57 (4.46)
Violating Human Rights × U.S. Military Alliance		-9.78 (6.05)	-8.19 (6.31)
Male			-5.15 (3.18)
Age			-0.03 (0.10)
Education			-20.84*** (5.10)
Income			0.53 (0.35)
(Intercept)	31.42*** (2.61)	29.02*** (3.00)	37.45*** (6.39)
Num. obs.	1000	1000	916

```

m2_ces <- lm(attack ~ alliance * hmrts, data = df_ces)
m3_ces <- lm(attack ~ alliance*hmrts +
             male + age_cat + edu4 + inc_10k, data = df_ces)

texreg(l = list(m1_ces, m2_ces, m3_ces),
       reorder.coef= c(2, 3, 4, 5, 6, 7, 8, 1),
       custom.coef.names = c("(Intercept)", "Violating Human Rights", "U.S. Military Alliance",
                              "Violating Human Rights $\\times$ U.S. Military Alliance",
                              "Male", "Age", "Education", "Income"),
       stars = c(0.01, 0.05, 0.1),
       digits = 2,
       caption = "Linear Estimate of Public Approval for Attacking the Third Country (CES Survey)",
       caption.above = T,
       include.ci = F,
       include.rmse = F,
       include.rsq = F,
       include.adjrs = F,
       label = "",
       custom.note = "",
       fontsize = "small") %>%
gsub(".begin.center.", "\\centering", .) %>%
gsub(".end.center.", "", .)

#### Figure S10: Impact of Treatments on Support for War (Binary Dependent Variable, YouGov Survey)

### Plot (10a) Support for Attack
# calculate the mean support for attack in each treatment group
df_ate_ces <- df_ces %>%

```

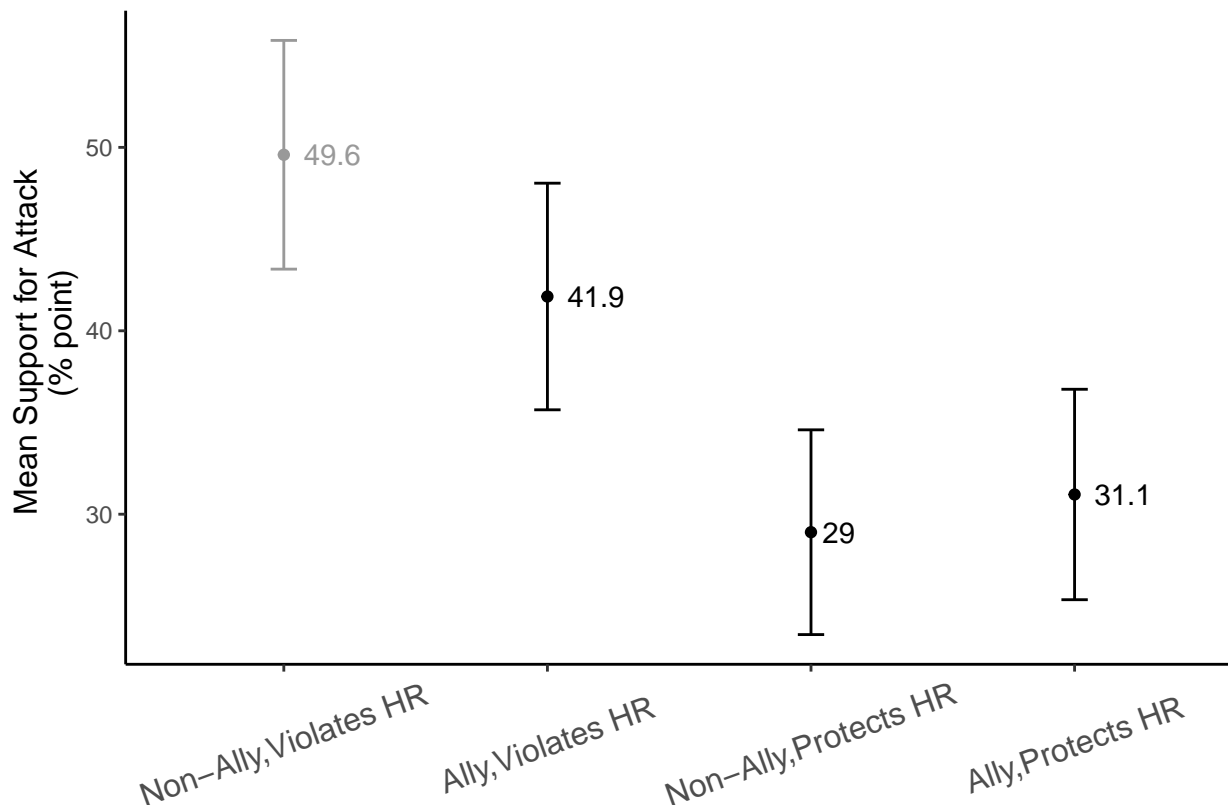
```

group_by(exp_4) %>%
  summarise(ate = mean(attack, na.rm = TRUE),
            n = n(),
            se = sd(attack, na.rm = TRUE) / sqrt(n)) %>%
  mutate(ci_low = ate - 1.96*se,
         ci_high = ate + 1.96*se)

# plot the mean support for attack in each treatment group and the 95% CI
p <- ggplot(df_ate_ces, aes(x = factor(exp_4, level=c('1', '3', '2', '4')), y = ate,
                           color=factor(exp_4))) +

  theme_classic() +
  geom_point()+
  geom_errorbar(aes(ymin=ci_low, ymax=ci_high), width=.1,
               position=position_dodge(0.05)) +
  scale_x_discrete(labels= c('Non-Ally, Violates HR', 'Ally, Violates HR',
                             'Non-Ally, Protects HR', 'Ally, Protects HR')) +
  scale_color_manual(values=c('#999999', 'black', 'black', 'black')) +
  theme(legend.position = "none") +
  labs(x = "", y = "Mean Support for Attack \n (% point)", size = 12) +
  geom_text(aes(label=round(ate, 1)), position=position_dodge(width=0.9),
           vjust=.5, hjust = -.35) +
  theme(axis.text.x = element_text(angle = 20, hjust = 0.5, vjust = 0.5, size = 12),
        axis.title.y = element_text(size=12))
p

```



```

# ggsave("ate-ces.pdf", width = 6, height = 4)

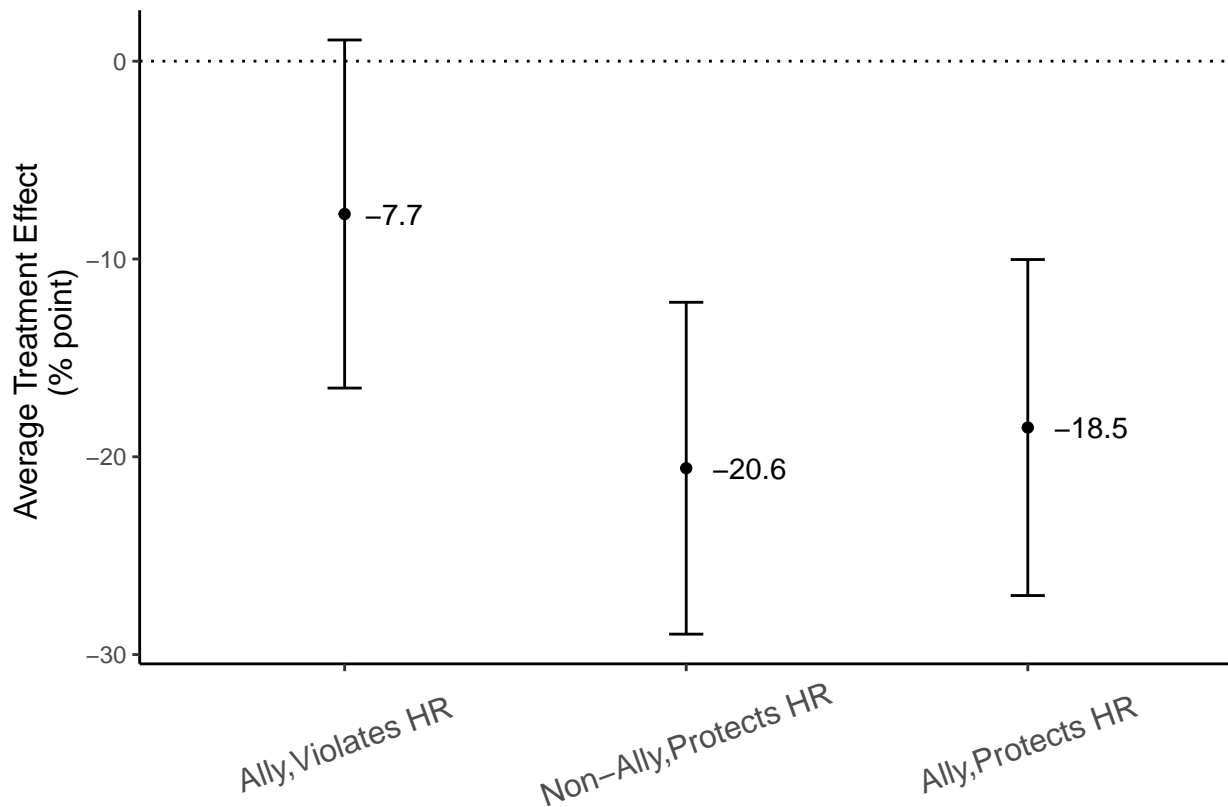
### Plot (10b) Average Treatment Effect on Support for Attack
# calculate the difference in support for attack between 2-4 against baseline condition
est <- rep(NA, 4)
ci_low <- rep(NA, 4)
ci_high <- rep(NA, 4)
se <- rep(NA, 4)

for(i in 2:4){
  test <- t.test(df_ces$attack[df_ces$exp_4==i],
                df_ces$attack[df_ces$exp_4==1])
  est[i] <- test[["estimate"]][["mean of x"]] - test[["estimate"]][["mean of y"]]
  ci_low[i] <- test[["conf.int"]][1]
  ci_high[i] <- test[["conf.int"]][2]
  se[i] <- test[["stderr"]]
}

df_ate_diff_ces <- data.frame(exp_4 = df_ate_ces$exp_4, est, ci_low, ci_high, se)
df_ate_diff_ces <- df_ate_diff_ces[-1, ]

# plot the differences and 95% CI
p1 <- ggplot(df_ate_diff_ces, aes(x = factor(exp_4, level=c('3', '2', '4')), y = est)) +
  theme_classic() +
  geom_point()+
  geom_errorbar(aes(ymin=ci_low, ymax=ci_high), width=.1,
                position=position_dodge(0.05)) +
  scale_x_discrete(labels= c('Ally, Violates HR', 'Non-Ally, Protects HR',
                             'Ally, Protects HR')) +
  labs(x = "", y = "Average Treatment Effect \n (% point)", size = 12) +
  geom_text(aes(label=round(est, 1)), position=position_dodge(width=0.9),
            vjust=.5, hjust = -.35) +
  geom_hline(yintercept = 0, linetype="dotted") +
  theme(axis.text.x = element_text(angle = 20, hjust = 0.5, vjust = 0.5, size = 12),
        axis.title.y = element_text(size=12))
p1

```



```
# ggsave("ate-diff-ces.pdf", width = 6, height = 4)
```

```
#### Table S11: Means of Public Support for War (Binary Dependent Variable, YouGov survey) ####
```

```
# group_mean of pre-election survey
```

```
mean_pre <- df_ces %>%
  group_by(exp_4) %>%
  filter(complete.cases(attack)) %>%
  summarise(mean = mean(attack, na.rm = TRUE),
            n = n())
```

```
mean_pre
```

```
## # A tibble: 4 x 3
##   exp_4 mean    n
##   <dbl> <dbl> <int>
## 1     1  49.6  248
## 2     2  29.0  255
## 3     3  41.9  246
## 4     4  31.1  251
```

```
# group mean of post-election survey
```

```
mean_post <- df_ces_post %>%
  group_by(exp_4) %>%
  filter(complete.cases(attack_post)) %>%
  summarise(mean = mean(attack_post, na.rm = TRUE),
            n = n())
```

```
mean_post
```

```

## # A tibble: 4 x 3
##   exp_4 mean    n
##   <dbl> <dbl> <int>
## 1     1  26.3  205
## 2     2  17.6  222
## 3     3  25.7  210
## 4     4  11.5  200

# difference of means
mean_post$mean - mean_pre$mean

## [1] -23.25531 -11.45204 -16.15563 -19.57570

# t.test by treatment group to obtain p-values
lapply(split(df_ces_post, factor(df_ces_post$exp_4)),
        function(x)t.test(data=x, attack_post ~ attack))

## $`1`
##
## Welch Two Sample t-test
##
## data:  attack_post by attack
## t = -5.4411, df = 178.45, p-value = 1.728e-07
## alternative hypothesis: true difference in means between group 0 and group 100 is not equal
## 95 percent confidence interval:
## -42.13038 -19.70445
## sample estimates:
## mean in group 0 mean in group 100
## 10.20408 41.12150
##
##
## $`2`
##
## Welch Two Sample t-test
##
## data:  attack_post by attack
## t = -4.864, df = 73.501, p-value = 6.367e-06
## alternative hypothesis: true difference in means between group 0 and group 100 is not equal
## 95 percent confidence interval:
## -46.55494 -19.49444
## sample estimates:
## mean in group 0 mean in group 100
## 8.641975 41.666667
##
##
## $`3`
##
## Welch Two Sample t-test
##

```

```

## data:  attack_post by attack
## t = -6.4816, df = 125.98, p-value = 1.86e-09
## alternative hypothesis: true difference in means between group 0 and group 100 is not equal
## 95 percent confidence interval:
##  -51.11618 -27.20340
## sample estimates:
##  mean in group 0 mean in group 100
##           9.677419           48.837209
##
##
##  `$4`
##
##  Welch Two Sample t-test
##
## data:  attack_post by attack
## t = -3.5463, df = 73.22, p-value = 0.0006855
## alternative hypothesis: true difference in means between group 0 and group 100 is not equal
## 95 percent confidence interval:
##  -33.103467  -9.283607
## sample estimates:
##  mean in group 0 mean in group 100
##           5.035971           26.229508

```